

Another Bayesian Analysis of the Thermal Challenge Problem

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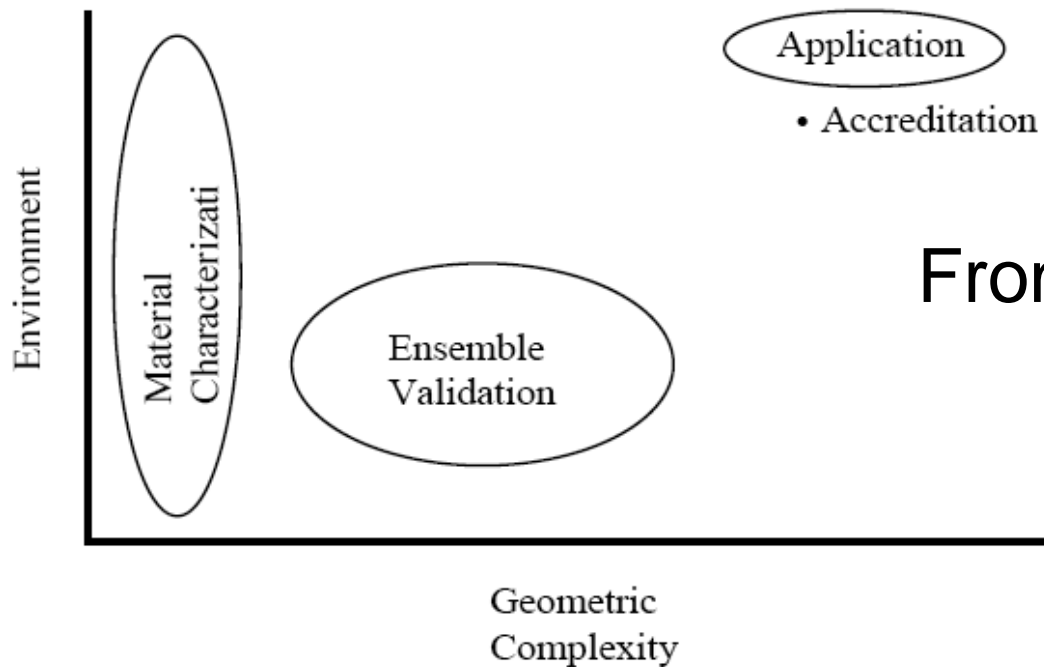
In this talk we describe the types of problems we work with at Los Alamos National Laboratory and how we see these problems fitting into the framework described in the tasking document for this workshop. We briefly describe tools and methods we have developed that utilize experimental data and detailed physics simulations for uncertainty quantification. Finally, we apply these tools to the thermal challenge problem. This statistical framework used here is based on the approach of Kennedy and O'Hagan (2001), but has been extended to deal with functional output of the simulation model.

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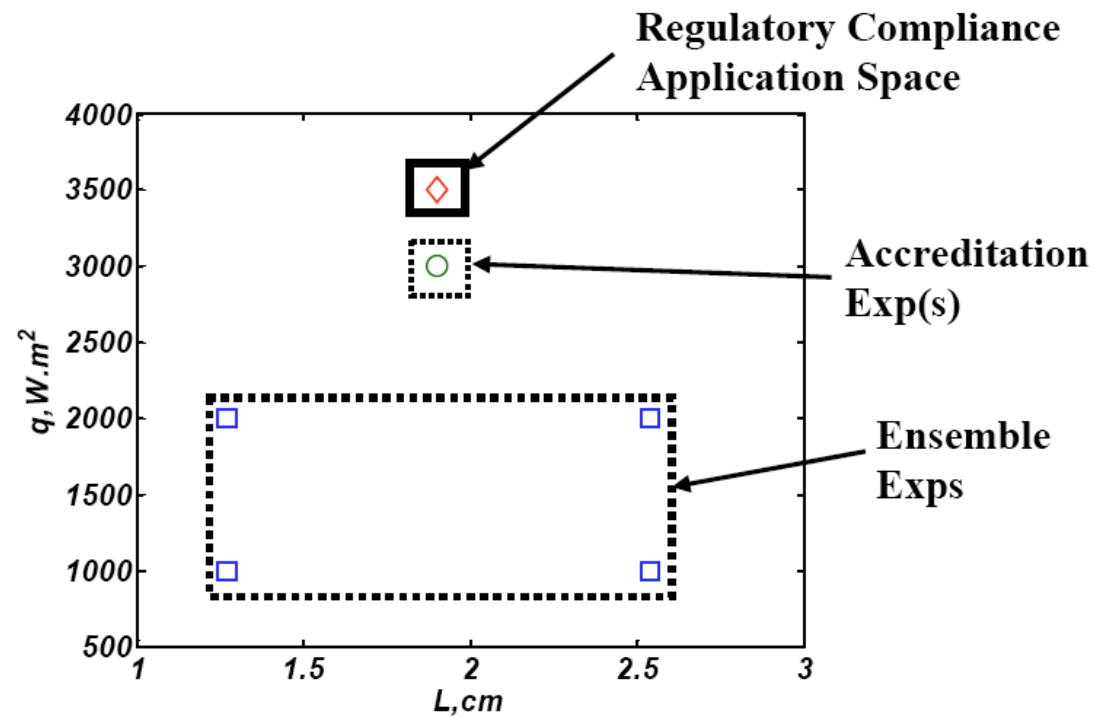
Thermal Validation Challenge

Problem: GPM solution

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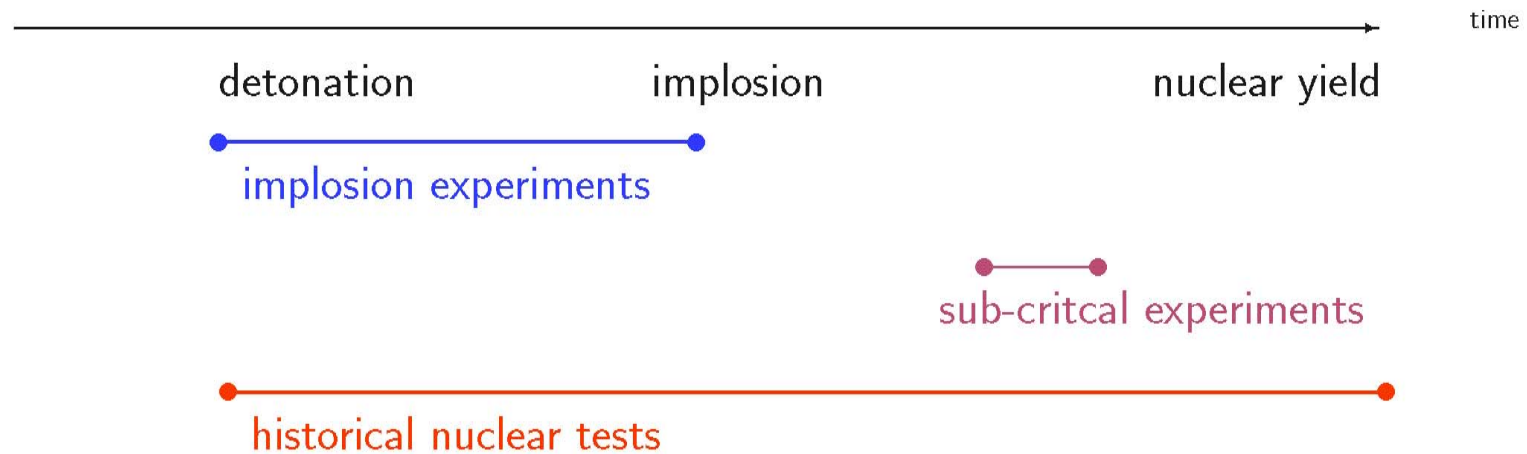
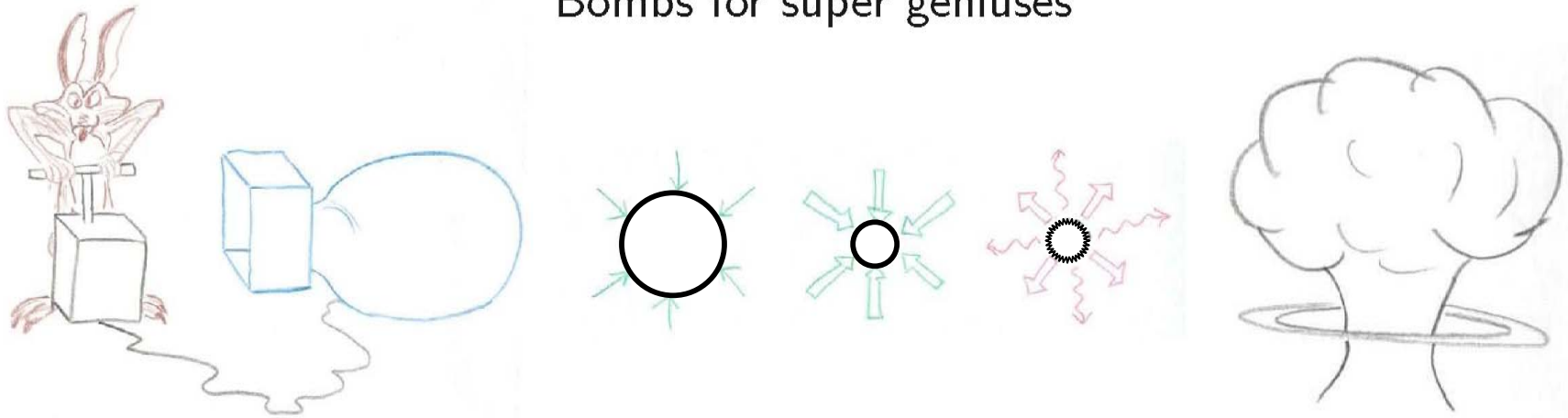


From the tasking document



Certification Issues at LANL

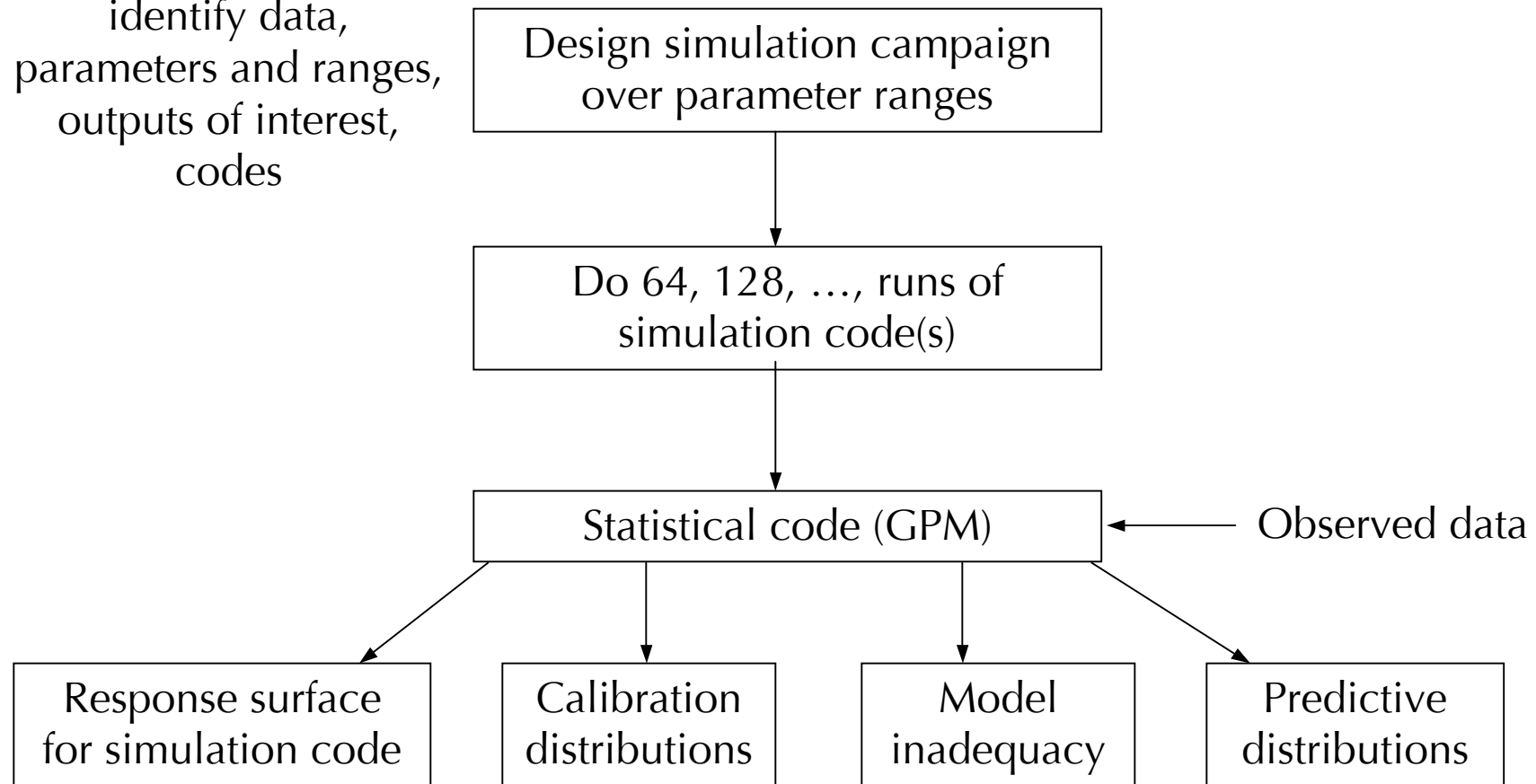
Bombs for super geniuses



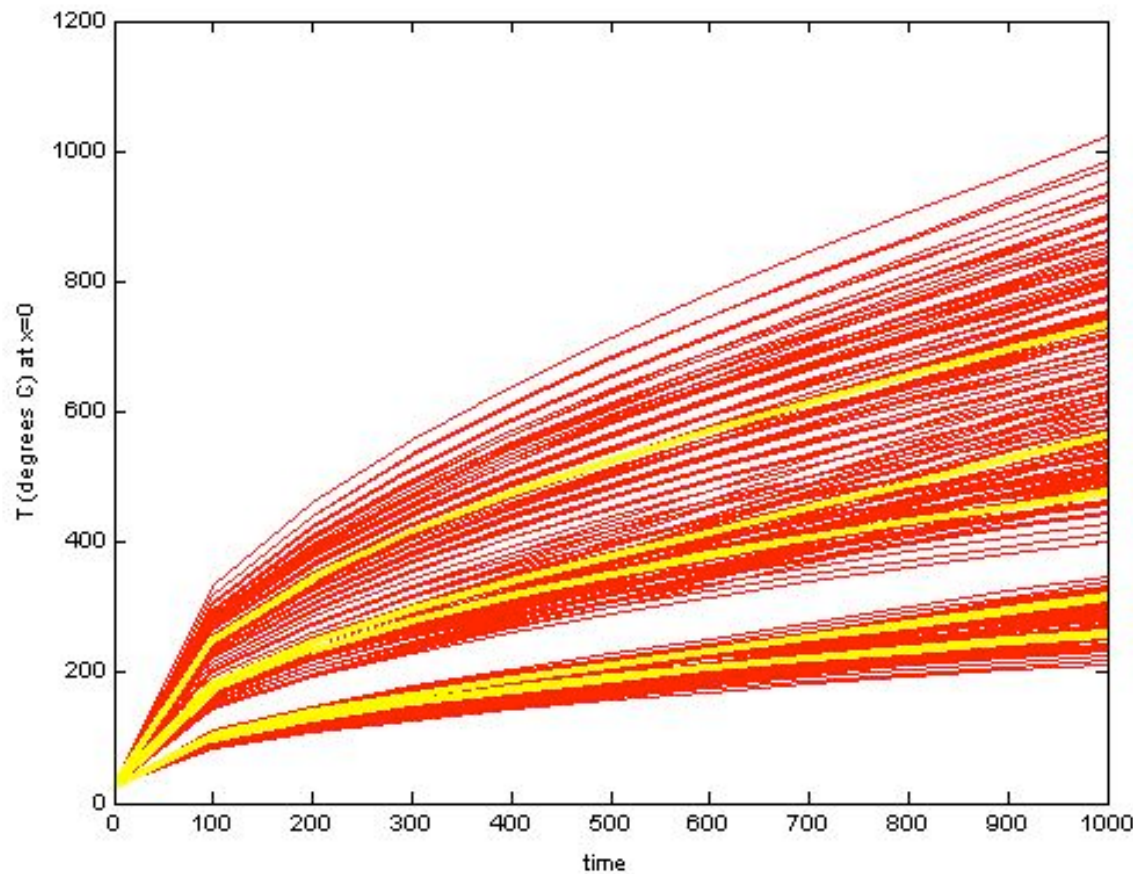
off-line experiments
materials, equations of state (EOS), high explosive (HE)

Our analyses use statistical methods to combine different simulation codes and observational data

Define problem:
identify data,
parameters and ranges,
outputs of interest,
codes



Uncalibrated sims and calibration data

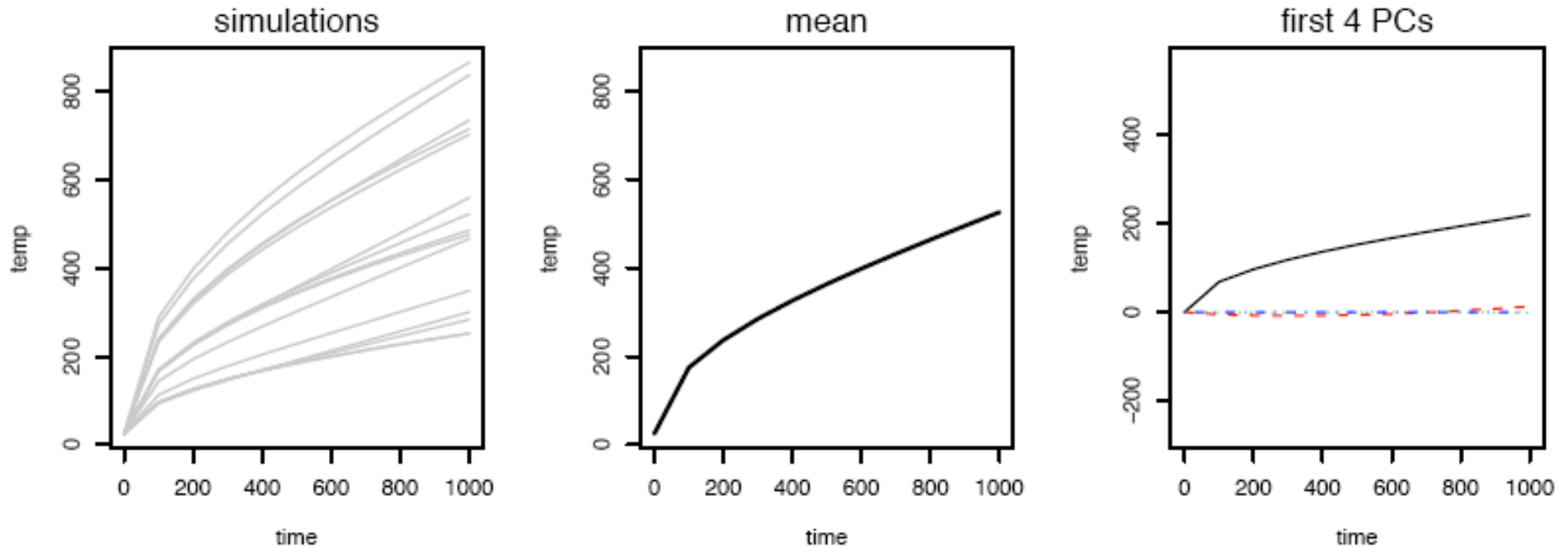


36 x 6 simulations

A single experiment
for each configuration
(1-5).

Basis representation of simulated profiles

The temperature profiles resulting from the 216 simulations are used to construct a mean-adjusted principal component representation.



Temperature profiles are represented as a function of the 4-d input parameters (x, θ) and PC basis functions $\phi_j(t)$:

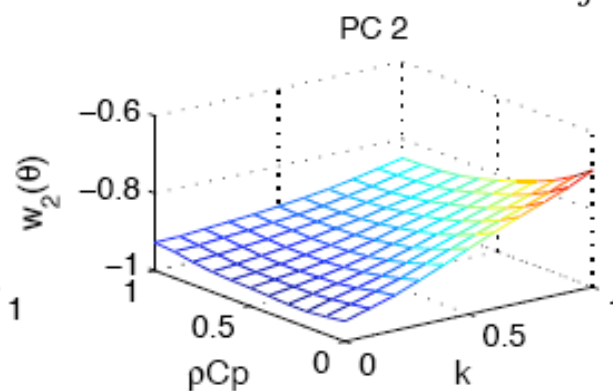
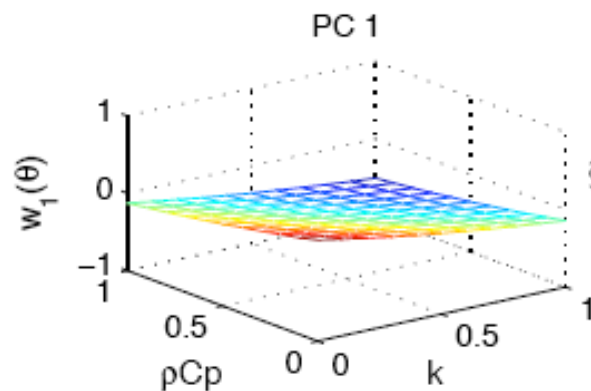
$$\hat{\eta}(x, \theta; k) = \sum_{j=1}^{p_{\eta}} w_j(x, \theta) \phi_j(t)$$

Gaussian process model to *emulate* simulation output

Gaussian process (GP) models are used to estimate the weights $w_j(\theta)$ at untried settings

$$\hat{\eta}(x, \theta; k) = \sum_{j=1}^{p_{\eta}} w_j(x, \theta) \phi_j(t)$$

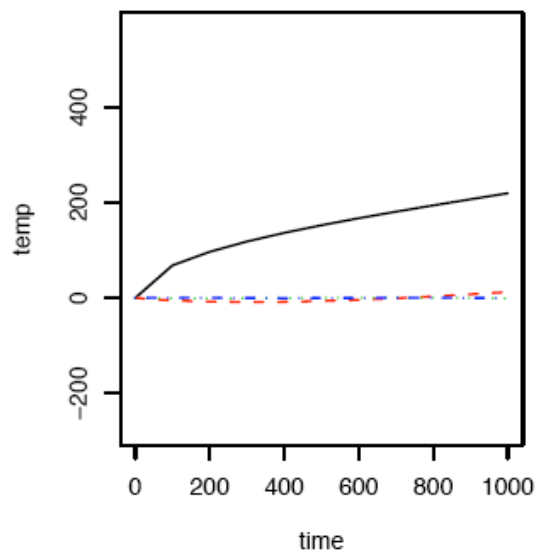
GP models $w_j(\mathbf{x}, \theta)$



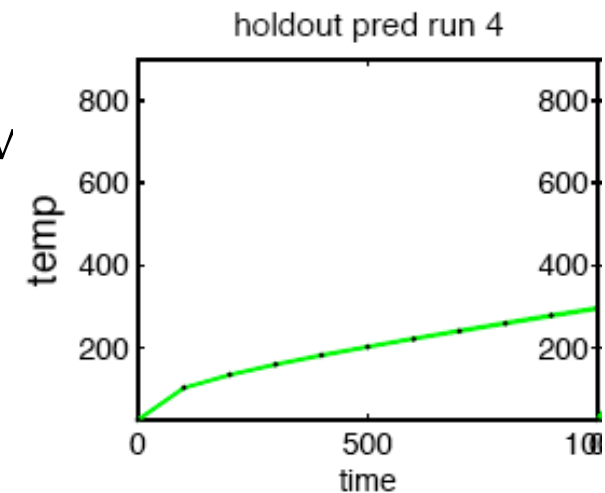
\mathbf{x} set to Config 4

first 4 PCs

Bases $\phi_j(\mathbf{x}, \theta)$

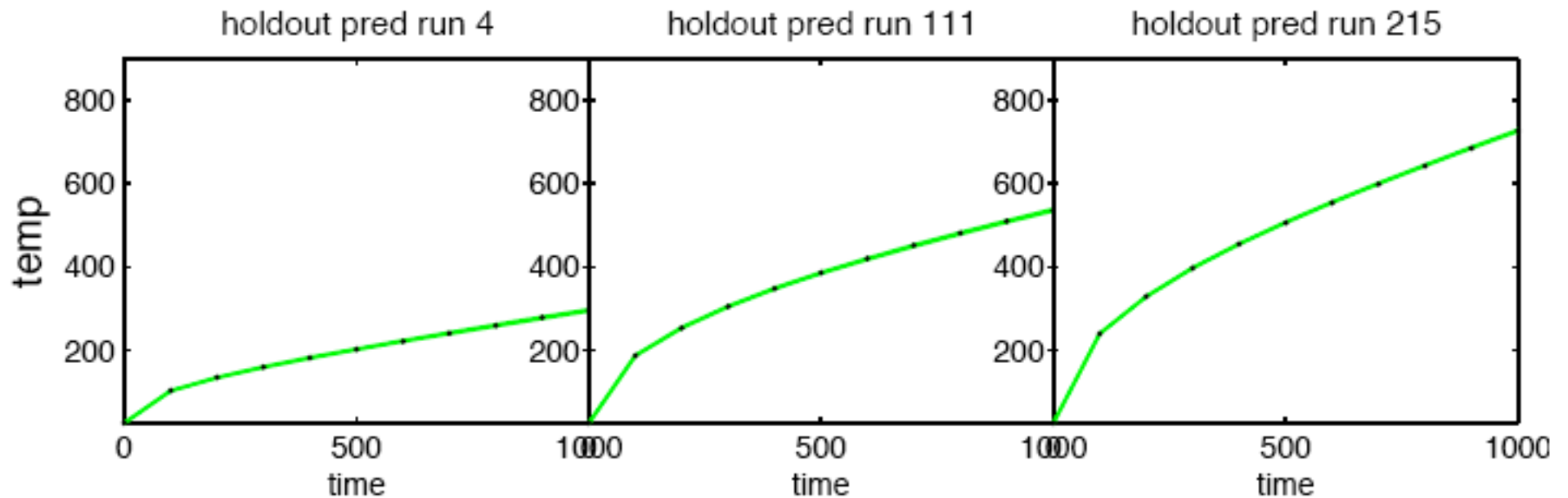


Prediction at new

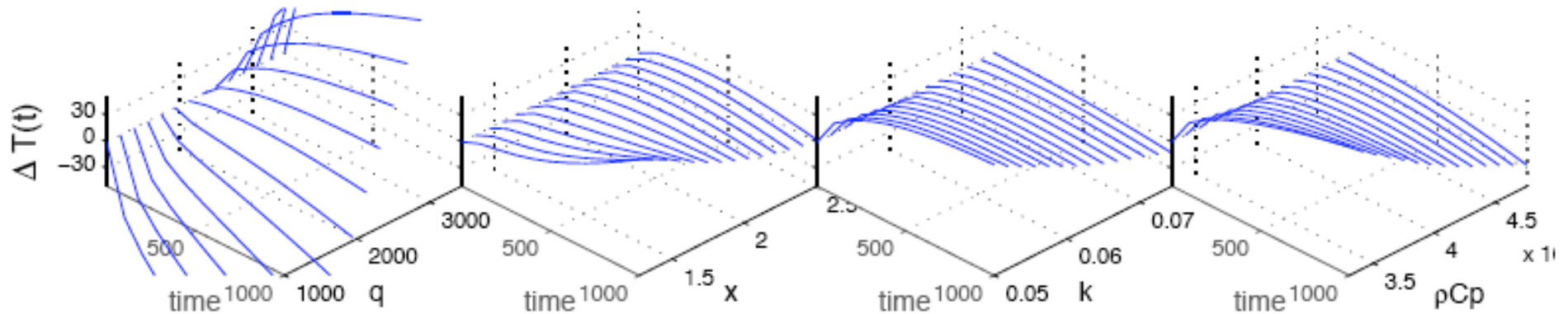


Checking the emulator

Errors ~ 0-3°C



Simulator emulation and sensitivity



Changes in the emulator prediction as each parameter is varied while holding the others at their midpoint.

Note: k and ρC_p have a similar effect on Temperature

Model of the data

$$y(t) = \eta(x, \theta; t) + \delta(x; t) + \epsilon_{rep}(t)$$

Posterior density:

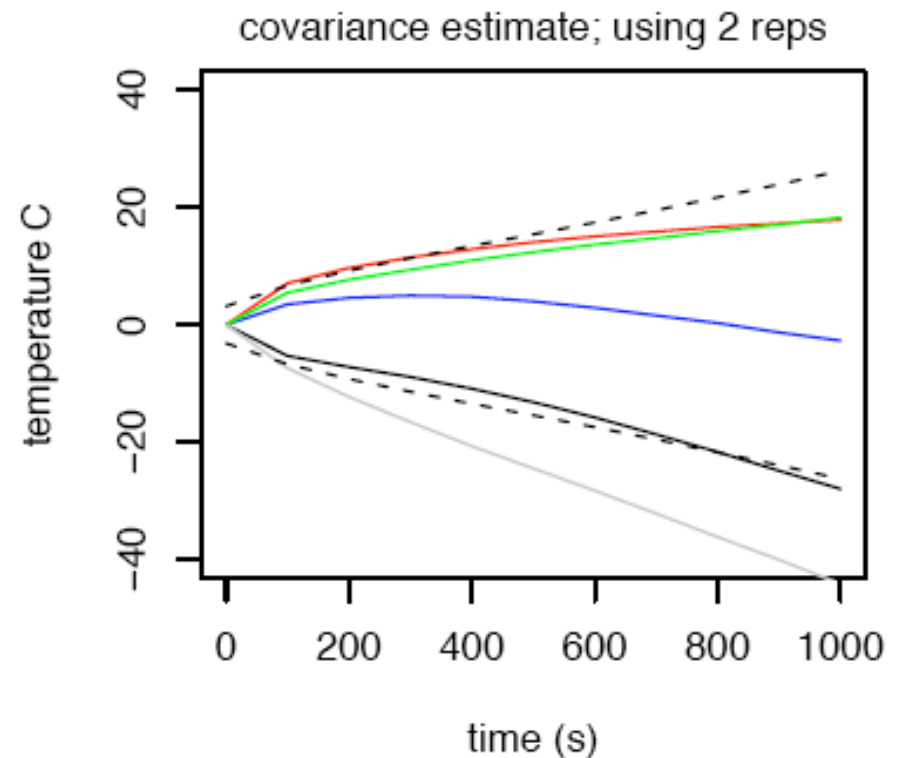
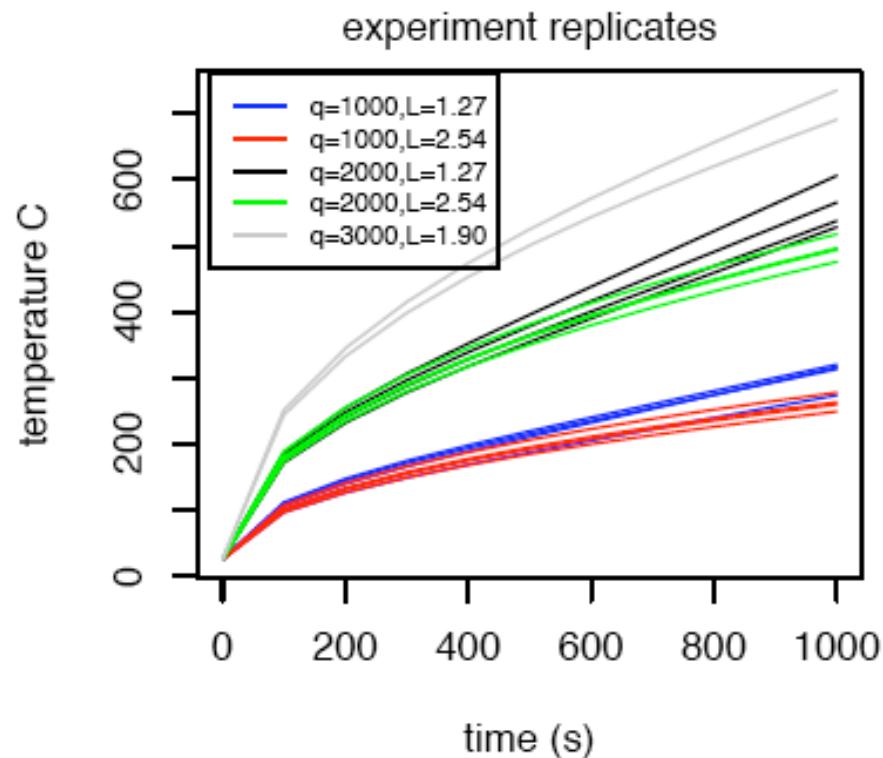
$$\begin{aligned} \pi(\delta(\cdot; t), \eta(\cdot, \cdot; t), \theta, \xi | y) &\propto L(y | \delta(\cdot; t), \eta(\cdot, \cdot, t), \theta, \Sigma_\epsilon) \\ &\times \pi(\delta(\cdot; t) | \xi) \times \pi(\eta(\cdot, \cdot; t) | \xi) \times \pi(\theta) \times \pi(\xi) \end{aligned}$$

Σ_ϵ is known, ξ controls statistical parameters governing $\eta(\cdot, \cdot; t)$, $\delta(\cdot; t)$, and Σ_ϵ .

Posterior for distributions/integrals computed via
MCMC

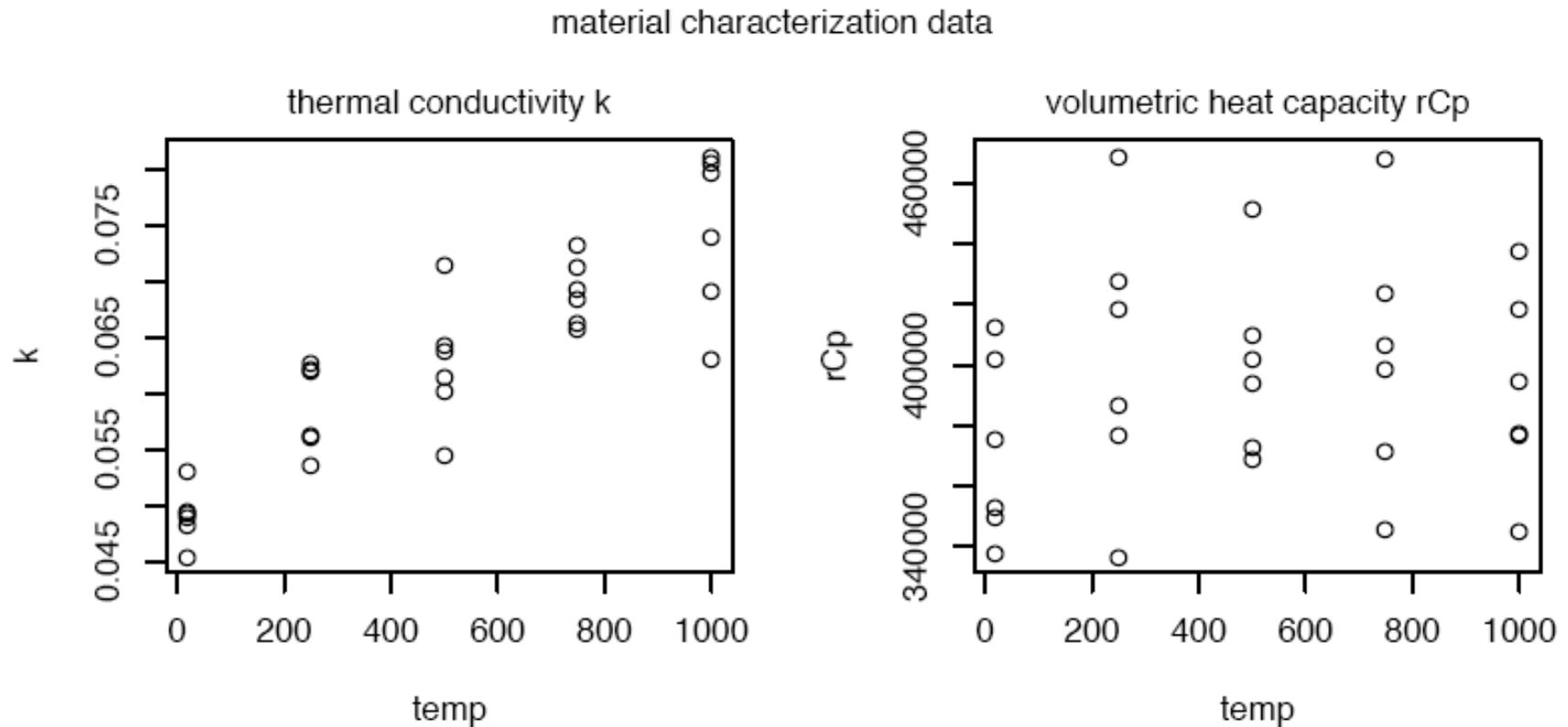
Modeling replicate variation

replicate variation



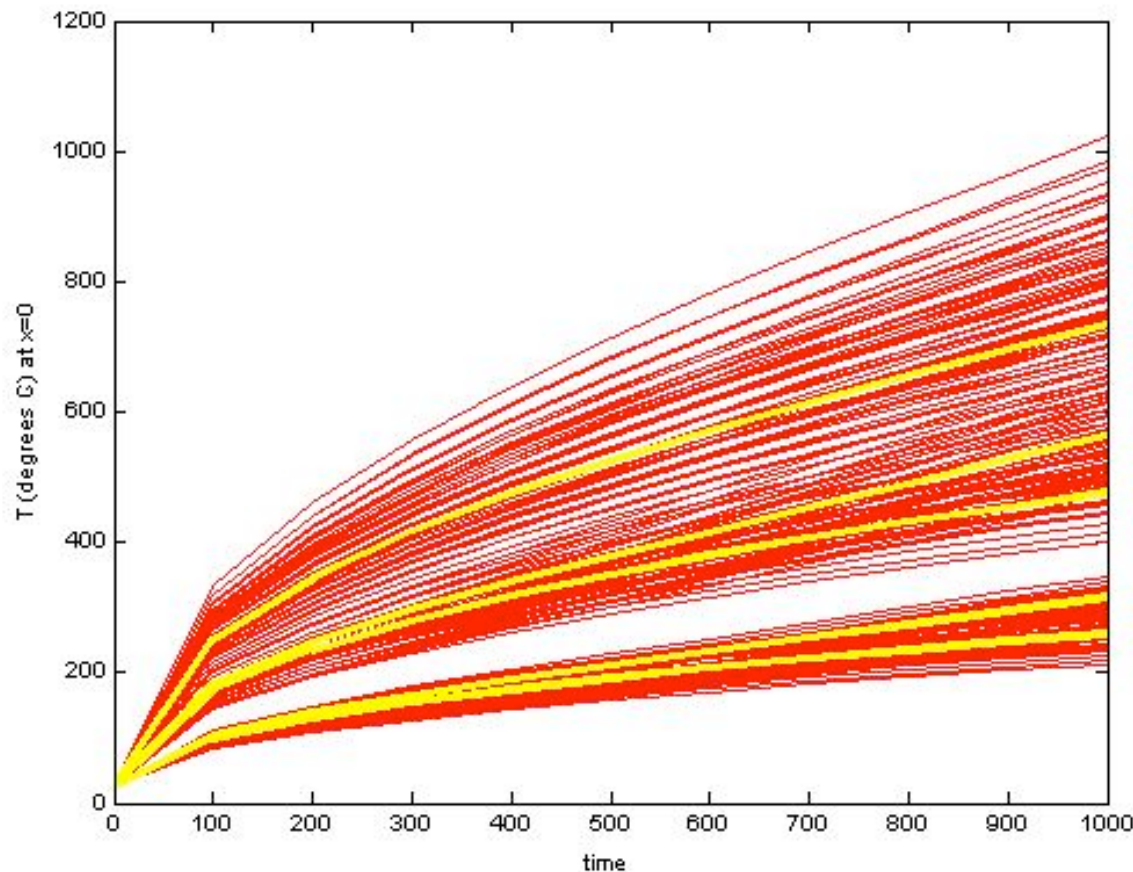
Correlated model for Σ_e

Use of material characterization data



Gives initial range for calibration

Using 1 experiment from 4 configurations

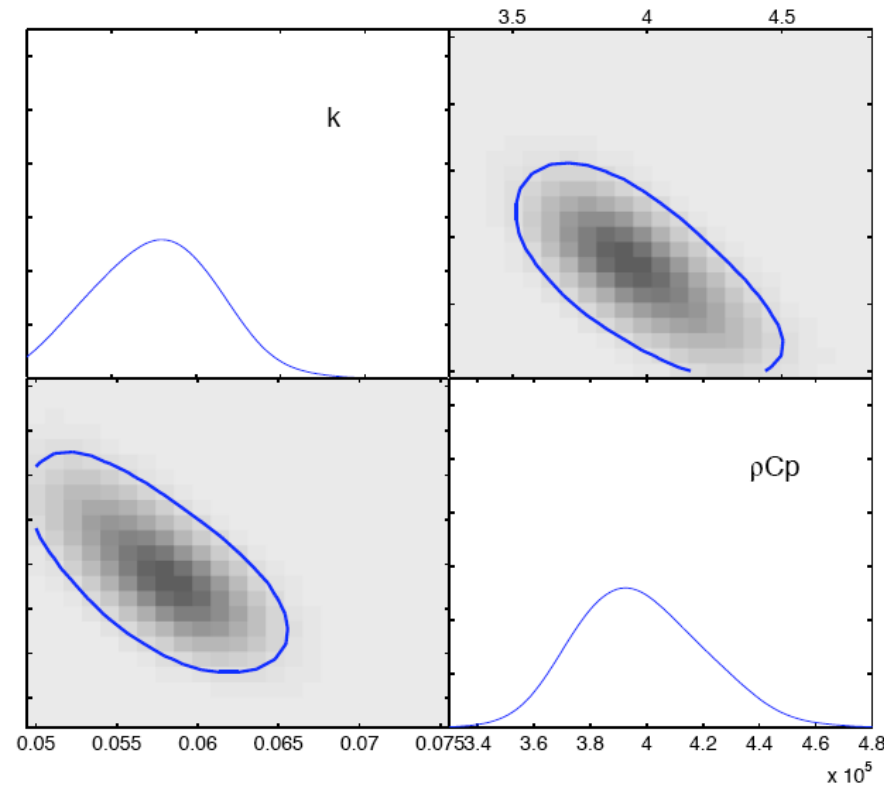


Similar to many LANL applications

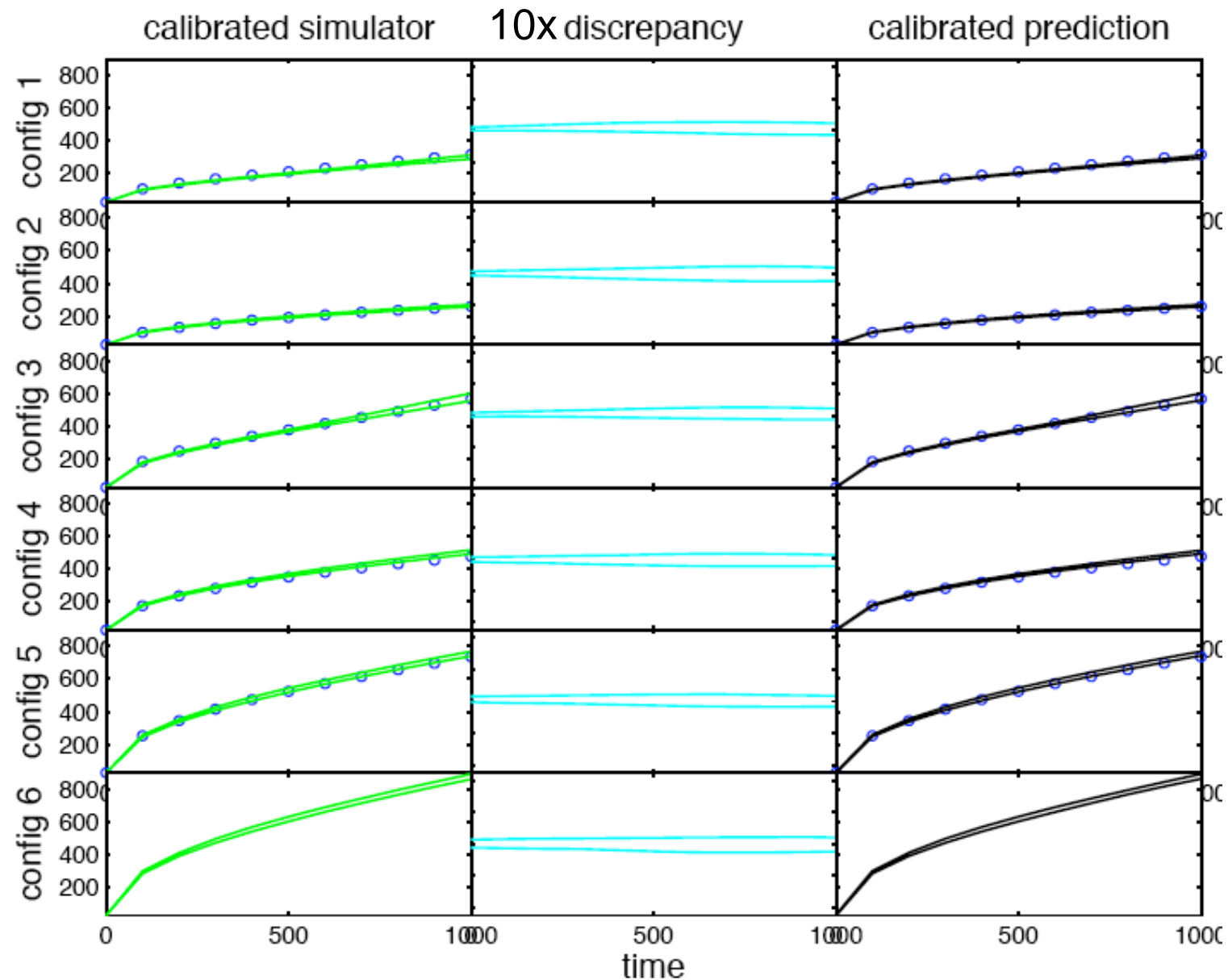
Can assess our prediction for the Accreditation experiment

Gives an idea of whether we can trust extrapolations

1st analysis: using only a single
expt from the 4 configurations

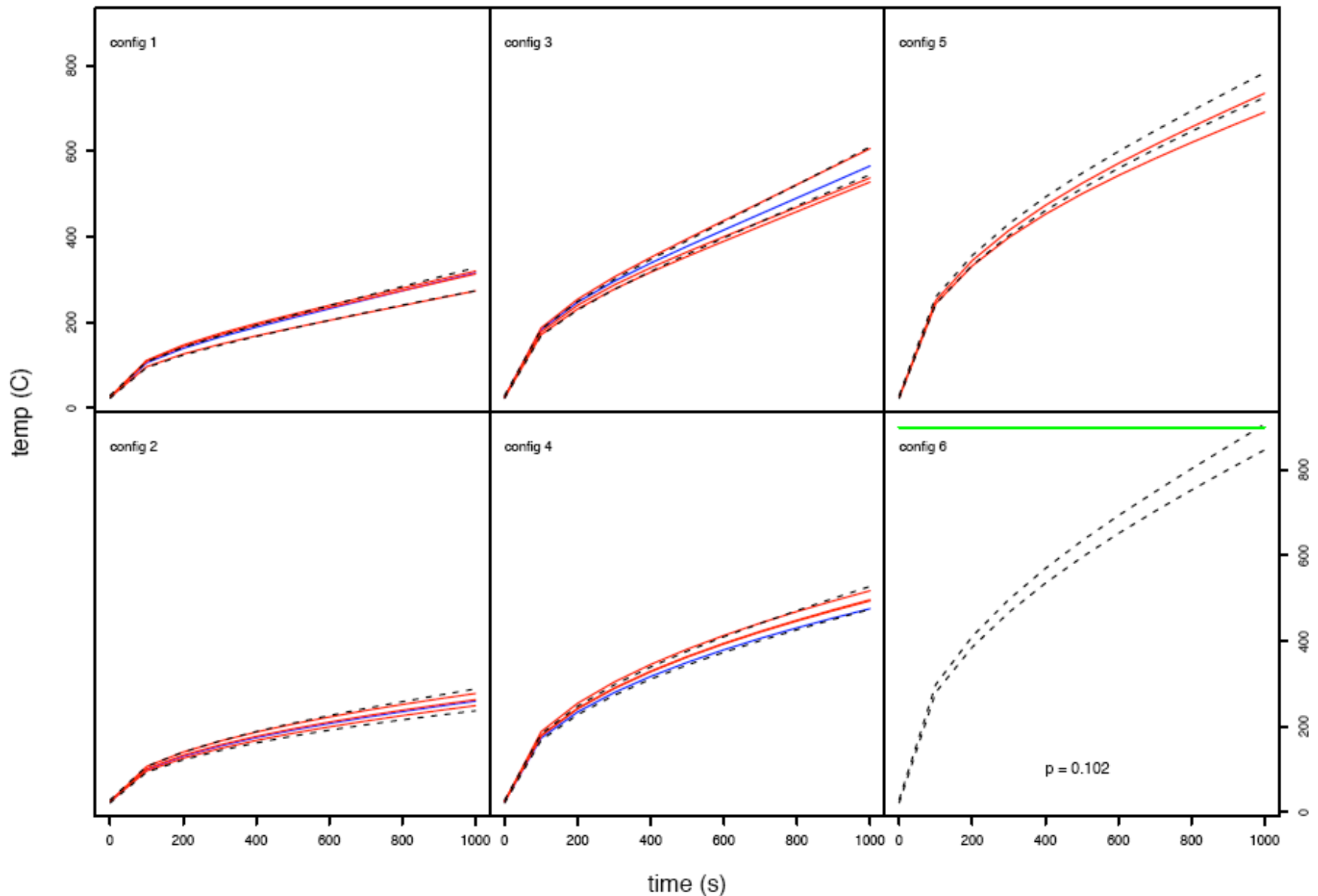


1st analysis: using only a single expt from the 4 configurations

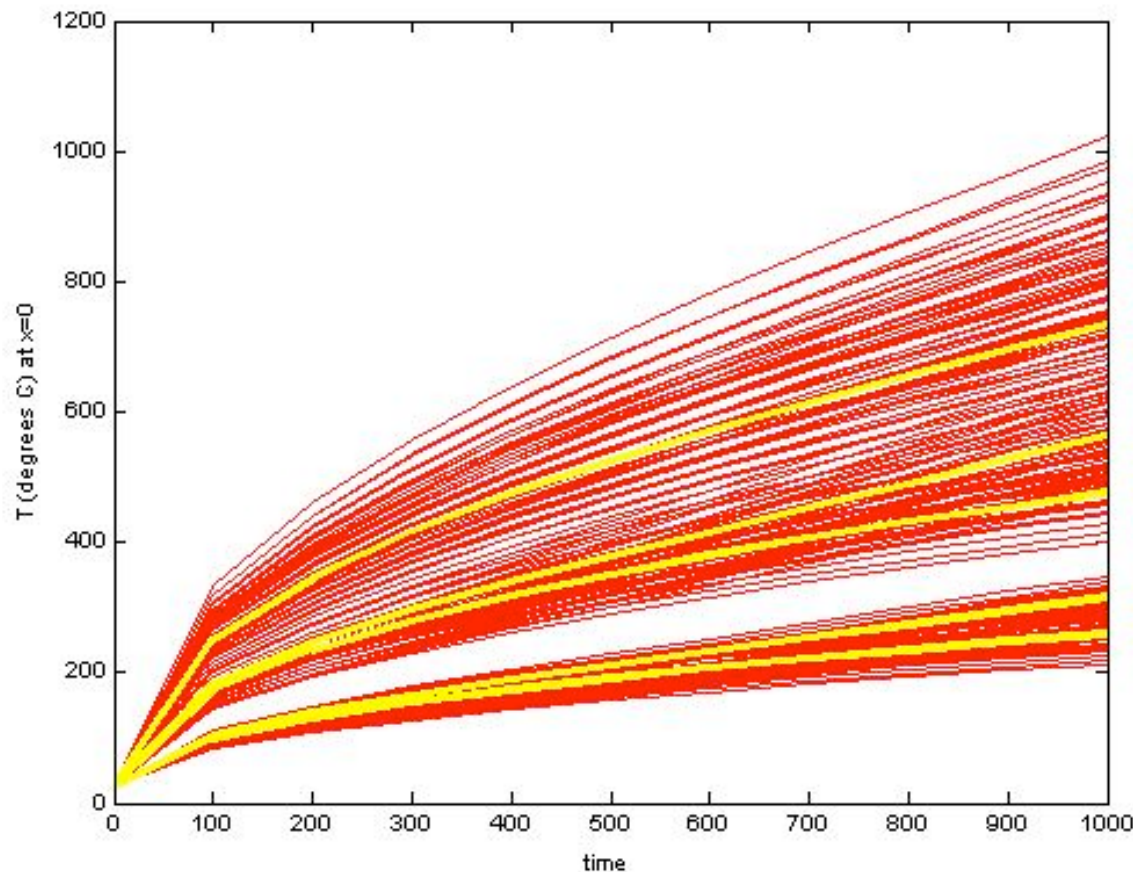


1st analysis: using only a single expt from the 4 configurations

Model predictions (90% intervals) using a single expt from configs 1–4



Using 1 experiment from 4 configurations + 1 accreditation expt

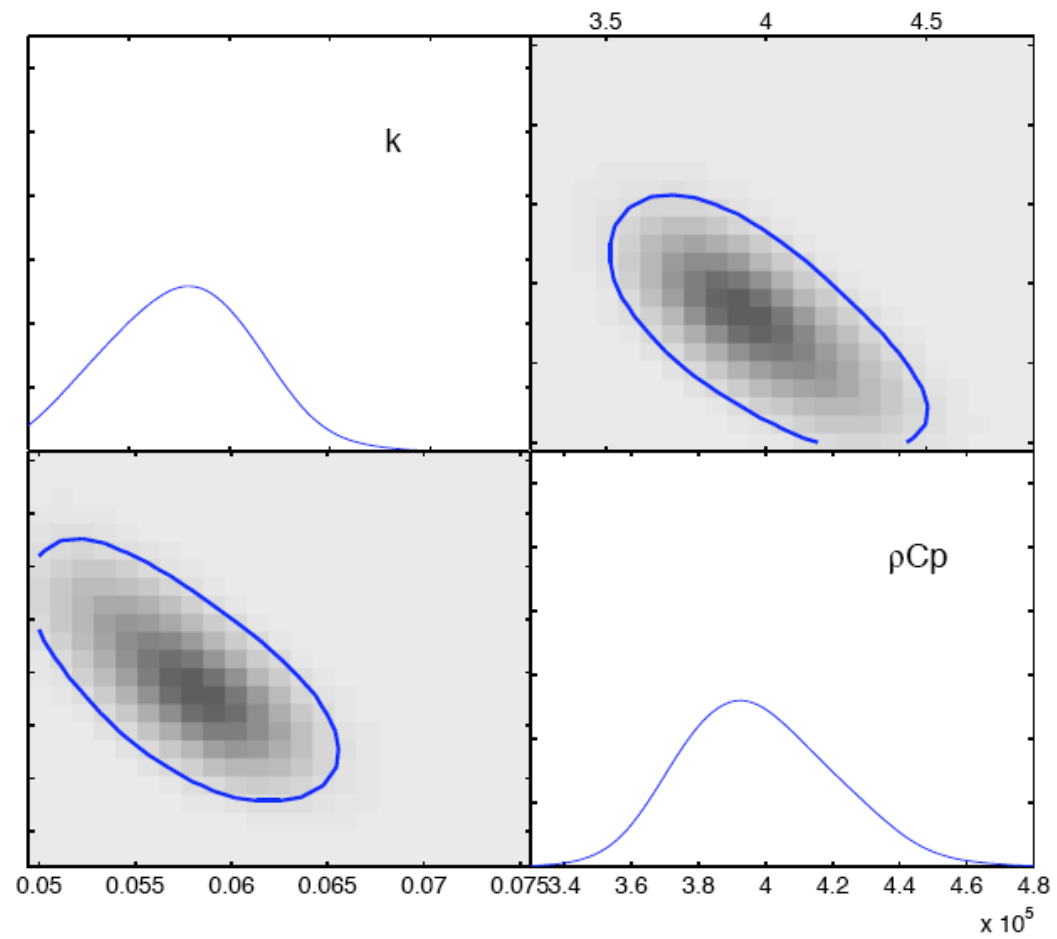


Similar to many LANL applications

Can assess our prediction for the Accreditation experiment

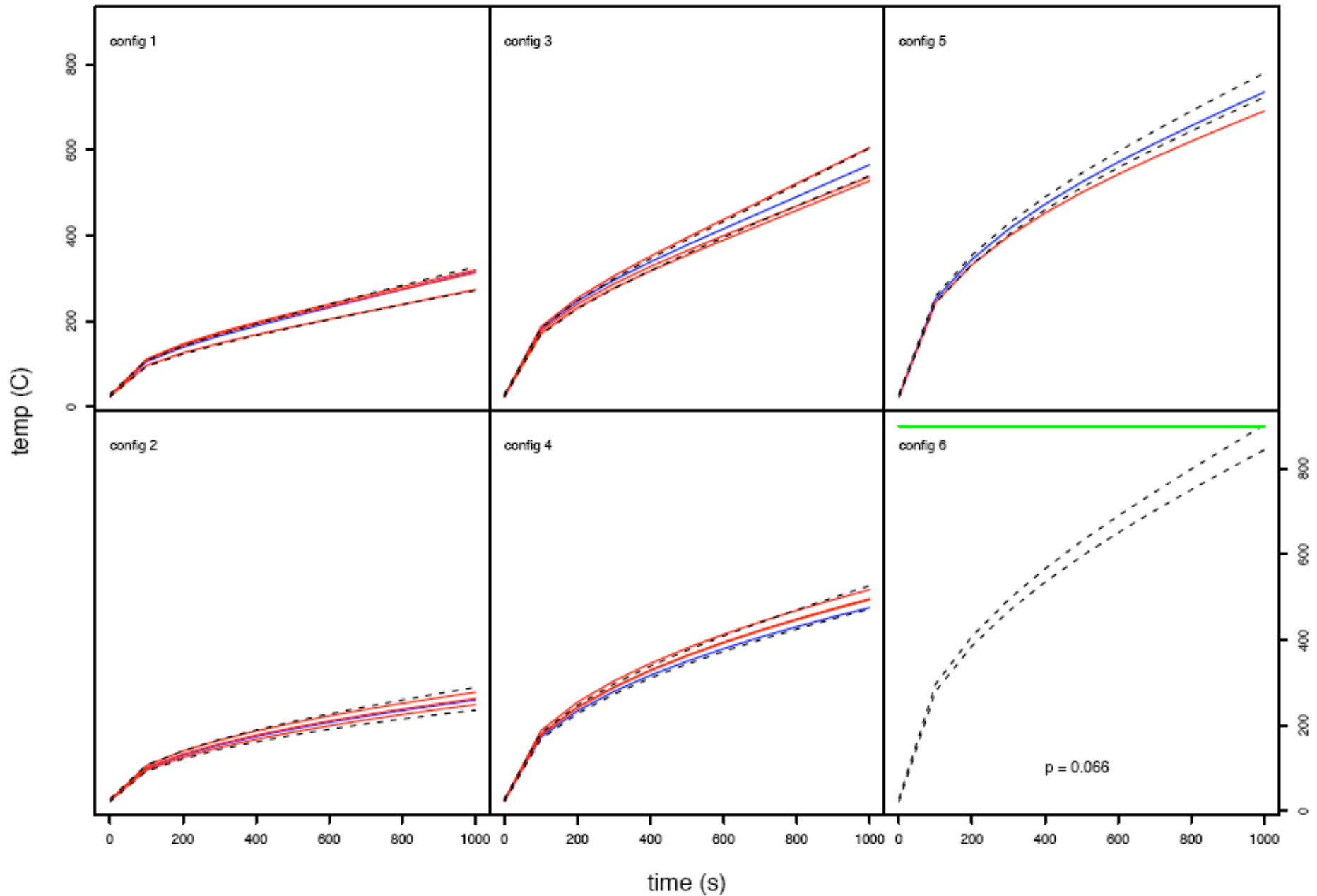
Gives an idea of whether we can trust extrapolations

2nd analysis: 4+1 expts



2nd analysis: 4 + 1 expts

Model predictions (90% intervals) using a single expt from configs 1–5



Discussion

- here, tail behavior of replicates important
- “reach” of constituent models in simulation models crucial for extrapolation
- Assessing trust in answers: experts, subject matter, holdout predictions, test statistics(?)
- Discrepancy model building with experts
- Metrics and discrepancy
- Focus on major sources of uncertainty and heartburn
- Philosophy: use all info at hand.
- Coupling multiple types of experiments